

# Detection of Cardiomyopathy using Support Vector Machine and Artificial Neural Network

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## ABSTRACT

With the growing needs of Medical diagnosis, recently several software tools and various algorithms have been proposed by the researchers for developing effective medical Expert systems. Diagnosing of heart disease like Cardiomyopathy is one of the important issue Cardiomyopathy refers to diseases of the heart muscle. These diseases enlarge the heart muscle or make it thicker and more rigid than normal. As cardiomyopathy worsens, the heart becomes weaker. It's less able to pump blood through the body and maintain a normal electrical rhythm. This can lead to heart failure or irregular heartbeats called arrhythmias which can be detected by analyzing ECG Signals. This paper describes aiming to develop an automated system for diagnosing of Cardiomyopathy using support vector machine and feed forward backpropagation technique. ECG data collected from Online PTB diagnostic ECG database ,preprocessing is applied for noise Cancellation and baseline correction then four time based features have been extracted and finally classification is been performed using Support vector machines and Artificial neural Networks. Results Shows that Feed forward backpropagation giving 98% accuracy for Healthy and 85% for Cardiomyopathy and Support Vector Machines gives 94.11 % accuracy for Healthy and 60 % accuracy for cardiomyopathy.

## General Terms

Biomedical Signal Processing, Pattern Recognition, Arrhythmias.

## Keywords

Cardiomyopathy, ECG, Support Vector Machines (SVM), Artificial Neural Networks (ANN), Feed forward back propagation Neural Network.

## 1. INTRODUCTION

Cardiomyopathy is a particular heart situation in which the Cardiac Muscle becomes inflamed and enlarged, due to which the heart muscle is stretched and becomes weak and hence it can't pump blood as fast as it should and if the heart muscle becomes too weak, it may develop heart failure. [1] There are many different types of Cardiomyopathy, caused by a range of factors from coronary heart disease to certain drugs. All these causes can lead to irregular heartbeats, heart failure, heart valve problems, and other complications. The Common symptoms of Cardiomyopathy are dyspnea (breathlessness) and peripheral edema (swelling of the legs). Cardiomyopathy is a significant clinical problem which is mainly generated by Volume/diastolic overload. To accommodate the increased blood volume, the heart chambers may stretch or dilate. Valvular regurgitation and congestive heart failure are two conditions that contribute to chamber dilation [2]. The person having cardiomyopathy shows irregular heartbeat called as Arrhythmias in Medical terminology. These can be detected

using ECG investigation .ECG refers to Electrocardiogram a noninvasive method in medical diagnosis it is the process of recording the electrical activity of the heart over a period of time using electrodes placed on a patient's body[3][4]. These electrodes detect the tiny electrical changes on the skin that arise from the heart muscle depolarizing during each heartbeat .In a conventional 12 lead ECG, ten electrodes are placed on the patient's limbs and on the surface of the chest. The overall magnitude of the heart's electrical potential is then measured from twelve different angles ("leads") and is recorded over a period of time (usually 10 seconds). In this way, the overall magnitude and direction of the heart's electrical depolarization is captured at each moment throughout the cardiac cycle. A typical ECG tracing is a repeating cycle of three electrical entities: a P wave (atrial depolarization), a QRS complex (ventricular depolarization) and a T wave (ventricular repolarization). The ECG is traditionally interpreted methodically in order to not miss any important findings. [3]

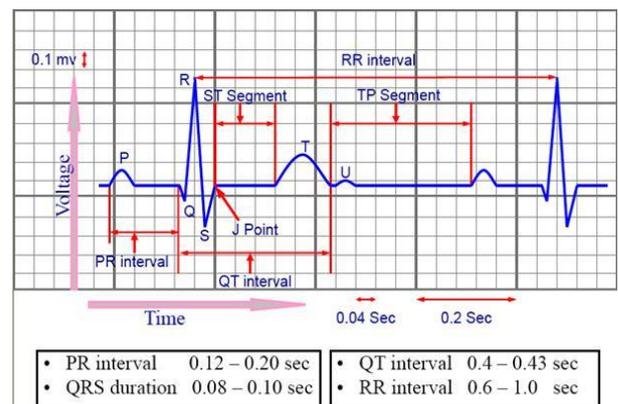


Figure 1: Normal Sinus Rhythm

The proposed system uses two techniques for classification Support Vector Machine and Feed forward backpropagation Neural Network the classification performance is Evaluated in terms of Accuracy, Sensitivity and positive predictive. The comparative study of Results shows better classification accuracy using Feed forward backpropagation Neural Network than Support Vector Machine. The paper is organized as follows, Section 2 describes Literature Review, Section 3 describes the details of Methodology and materials, Section 4 presents the Results and Discussion and final conclusion is given in section 5.

## 2. LITRATURE REVIEW

In an effort of automated detection of heart diseases various research and algorithms have been developed those are based on ECG data analysis and classification. Digital signal analysis, Fuzzy Logic methods, Artificial Neural Network, Hidden Markov Model, Genetic Algorithm, Support Vector Machines, Self-Organizing Map, Bayesian are various

classification techniques proposed and under Evaluation each technique is having its own advantages and disadvantages.

Mirela Ovreiu and Dan Simon (2012) shown that the clinical ECG data can be classified as Cardiomyopathy and Non Cardiomyopathy using neuro-fuzzy network training by biogeography-based optimization(BBO) and got 60 % accuracy on test data[5]. Ahmad Shukri.M.H et al (2012) proposed a method for detection of Cardiomyopathy with Elman Neural Network .In his approach he took total of 600 ECG beat samples from an established online database. Nine time-based descriptors from leads I, II and III were used for training, testing and validation of the network structures. A total of five hidden-node node structures were tested with four different learning algorithms. Results show that all the network structure managed to achieve more than 90% classification accuracy. The fastest convergence was achieved with the Levenberg-Marquardt algorithm with an average of 16 epochs [6]. a similar approach for detecting cardiomyopathy is proposed by Ali M.S.A.M (2012) they used Multilayered Perception Network for classification, they worked on PTB diagnostic ECG database which includes data for normal, cardiomyopathy and other arrhythmias .they trained total 600 beats and test five different MLP structures, four different algorithms were implemented to obtain the best classification accuracy and results showed that the Levenberg Marquardt Algorithm Gives Highest average Classification accuracy of 98.9%.[7]. A. T. SAYAD et al (2014) presented diagnostic system for Heart disease using data mining and artificial neural network (ANN) techniques. From the ANN, a Multilayer perceptron neural network along with back propagation algorithm is used to develop the system and works efficiently. [8]. V .K. Srivastava et al (2013) presented the system based on the application of Neuro Fuzzy network for reliable heartbeat classification on the basis of the ECG wave form. The potential of accuracy rate for Neuro Fuzzy system as ECG diagnostic decision aid is very high.[9]. Lai Khin Wee et al (2009) developed the automation data capturing and sending system based on digitized ECG image that enables the patient monitoring process with the medical devices without any wire connections. The output of the developed program is a self-generated report which including patients' information and heart rate. It will then be send through the internet connection to the user's recipients. [10] K.S.Kavitha et al (2010) proposed a system for detection of heart disease based on feed forward neural network architecture and genetic algorithm in their work they uses Genetic algorithm for training that Overcomes the problem of local minima with backpropagation algorithm.[11]. Mrs. B. Anuradha and V.C.Veera Reddy(2005) proposed arrhythmia Classification using Fuzzy Classifier they showed that The cardiac arrhythmia classification using fuzzy logic approach has given better performance results when compared to analytical method and using artificial neural networks for classification. More over problems associated with conventional neural network arch such as learning rate limitation and difficulty in selecting the optimal number of hidden units are eliminated [12].Anand Kumar Joshi et al (2014) in his review paper focused on analysis of ECG signals for Arrhythmia and abnormalities. And concluded that the DWT discrete wavelets transform analyzes the ECG signals and neural networks are the efficient classifier for arrhythmias [13].

### 3. METHODOLOGY AND MATERIALS



Figure 2: Proposed System Methodology

#### 3.1 Data Acquisition

The system works on total 63 subjects ECG data which is obtained from online PTB diagnostic ECG database The database contains 549 records from 290 subjects (aged 17 to 87, mean 57.2; 209 men, mean age 55.5, and 81 women, mean age 61.6; ages were not recorded for 1 female and 14 male subjects). Each subject is represented by one to five records [14]. The proposed work uses signals of 49 healthy subjects and 14 cardiomyopathy subjects.

#### 3.2 Preprocessing

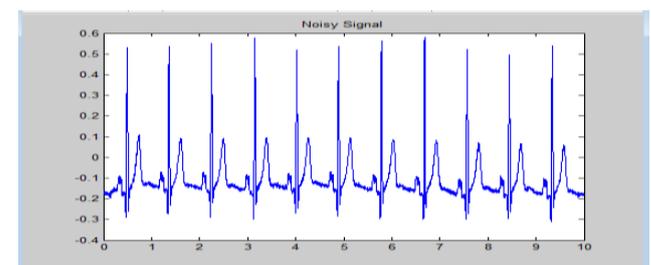
Preprocessing is the task of data preparation for feature extraction that refines the data by removing the unwanted artifacts. Two dominant artifacts present in ECG recordings are: (1) high-frequency noise caused by electromyogram induced noise, power line interferences, or mechanical forces acting on the electrodes; (2) baseline wander (BW) that may be due to respiration or the motion of the patients or the instruments. These artifacts severely limit the utility of recorded ECGs and thus need to be removed for better clinical evaluation.

In this method we had applied the Band pass filters with 0.5 to 45 Hz that is the cutoff low frequency to get rid of baseline wander and the cut off frequency to discard high frequency noise. Then Moving average Filters are applied for further enhancement.

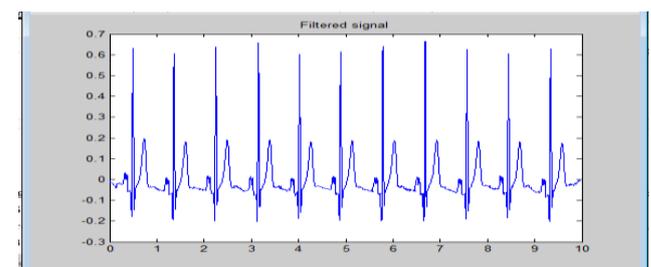
$$[i] = \frac{1}{M} \sum_{j=0}^{M-1} x(i + j) \text{ ----- (1)}$$

Equation of the moving average filter

In this equation,  $x [ ]$  is the input signal,  $y [ ]$  is the output signal, and  $M$  is the number of points used in the moving average. This equation only uses points on one side of the output sample being calculated. [15].



(a)Raw Signal



(b)preprocessed Signal

Figure 3: Raw and Preprocessed Signal

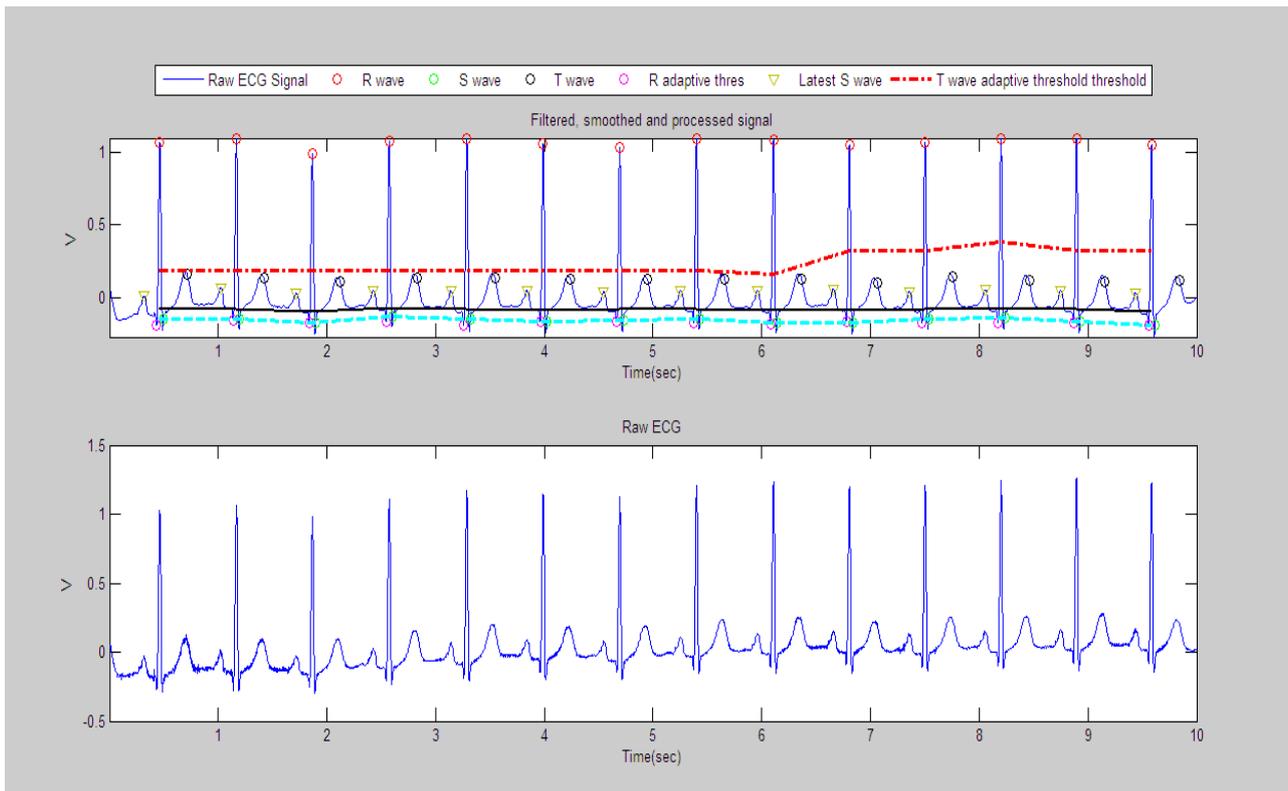
After filtering Different Quality measures Like MSE, PSNR and SNR are applied to judge the quality of signals for noise removal and baseline correction.

**Table 1: MSE, PSNR and SNR values of ECG Signals.**

Signal ID	MSE	PSNR	SNR
s0311rem(Healthy)	0.006272987	70.19006	1.01541
s0402rem(Healthy)	0.000469348	81.44985	0.589849
s0420_rem(Cardio)	0.000711506	79.64302	1.229599
s0456_rem(Cardio)	0.000553734	80.73179	0.282322

### 3.3 Feature Extraction

The extracted feature from the ECG signal plays a vital in diagnosis of Cardiomyopathy. The development of accurate and quick methods for automatic ECG feature extraction is of major importance. One cardiac cycle in an ECG signal consists of the P-QRS-T waves. The duration of the waves and the time Intervals of the Waves are of Importance to detect abnormality or Arrhythmias in an ECG. Detection of P wave, QRS complex wave is difficult as ECG signal varies with time due to psychological conditions and presence of noise the proposed feature extraction scheme uses Adaptive thresholding method for detecting the p wave, QRS complex and T waves of an ECG signal [16].



**Figure 4 : Signal with P –QRS complex and T waves Detected**

Once the peaks are detected, it is possible to calculate the subsequent intervals of the waves for detecting arrhythmias.

In this study The Four Time based features PR interval, QT interval, RR interval and QRS interval are calculated.

**Table 2: Normal and Calculated range of ECG wave Intervals**

Sample No	PR Interval	QT Interval	RR Interval	QRS Interval	Disease Detected
<b>Normal Range</b>	<b>0.12 to 0.20 s</b>	<b>0.35 to 0.44 s</b>	<b>0.6 to 1.0 s</b>	<b>0.08 to 1.0s</b>	<b>Healthy</b>
2	0.083636	0.378336	0.9218	0.129364	Cardiomyopathy
3	0.100167	0.436942	0.853091	0.087583	Cardiomyopathy
4	0.089667	0.193913	0.841667	0.069083	Cardiomyopathy
5	0.04	0.37819	0.725071	0.144929	Cardiomyopathy
11	0.1462	0.503417	0.9721	0.0932	Cardiomyopathy
34	0.129	0.335154	1.0345	0.097	Healthy
35	0.1161	1.383794	1.034778	0.1025	Healthy
36	0.144909	0.324022	0.908545	0.059455	Healthy

37	0.138	0.261558	0.9595	0.06	Healthy
38	0.142	0.351099	0.9744	0.085727	Healthy
39	0.103615	0.337534	0.770923	0.052923	Healthy
40	0.107231	0.410371	0.785154	0.080462	Healthy
41	0.118615	0.269958	0.782154	0.055538	Healthy
42	0.137091	0.322694	0.967364	0.089273	Healthy
43	0.113615	0.391447	0.781692	0.082231	Healthy

### 3.4 Classification

The Extracted Features PR, RR, QT and QRS intervals are normalized before classification, two classification methods Support vector Machines and Feed forward backpropagation neural networks are applied and finally the performance is measured using some statistical metrics.

#### 3.4.1 Support Vector Machines

Support vector machine (SVM) is a novel learning Machine introduced first by Vapnik [17] Support Vector Machine is based on the concept of decision planes that define decision boundaries. A decision plane is one that separates between a set of objects having different class memberships. SVM supports both regression and classification tasks and can handle multiple continuous and categorical variables. [18]. in this study we had used Linear SVM as we are having only two classes. Given some training data  $\mathcal{D}$ , a set of  $n$  points of the form

$$\mathcal{D} = \{(\mathbf{x}_i, y_i) \mid \mathbf{x}_i \in \mathbb{R}^p, y_i \in \{-1, 1\}\}_{i=1}^n$$

Where the  $y_i$  is either 1 or -1, indicating the class to which the point  $\mathbf{x}_i$  belongs. Each  $\mathbf{x}_i$  is a  $p$ -dimensional real vector. We want to find the maximum-margin hyperplane that divides the points having  $y_i = 1$  from those having  $y_i = -1$ . Any hyperplane can be written as the set of points  $\mathbf{x}$  satisfying

$$\mathbf{w} \cdot \mathbf{x} - b = 0,$$

Where  $\cdot$  denotes the dot product and the (not necessarily normalized) normal vector to the hyperplane. The parameter  $\frac{b}{\|\mathbf{w}\|}$  determines the offset of the hyperplane from the origin along the normal vector  $\mathbf{w}$ . If the training data are linearly separable, we can select two hyperplanes in a way that they separate the data and there are no points between them, and then try to maximize their distance. The region bounded by

them is called "the margin". These hyperplanes can be described by the equations [19]

$$\mathbf{w} \cdot \mathbf{x} - b = 1$$

and

$$\mathbf{w} \cdot \mathbf{x} - b = -1.$$

Geometrically, the distance between these two hyperplane is  $\frac{2}{\|\mathbf{w}\|}$ , so to maximize the distance between the planes we want to minimize  $\|\mathbf{w}\|$ . As we also have to prevent data points from falling into the margin, we add the following constraint: for each  $i$  either

$$\mathbf{w} \cdot \mathbf{x}_i - b \geq 1 \quad \text{for } \mathbf{x}_i \text{ of the first class}$$

Or

$$\mathbf{w} \cdot \mathbf{x}_i - b \leq -1 \quad \text{for } \mathbf{x}_i \text{ of the second.}$$

This can be rewritten as:

$$y_i(\mathbf{w} \cdot \mathbf{x}_i - b) \geq 1, \quad \text{for all } 1 \leq i \leq n. \quad (1)$$

We can put this together to get the optimization problem:

Minimize (in  $\mathbf{w}, \mathbf{b}$ )

subject to (for any  $i = 1, \dots, n$ )

$$y_i(\mathbf{w} \cdot \mathbf{x}_i - b) \geq 1.$$

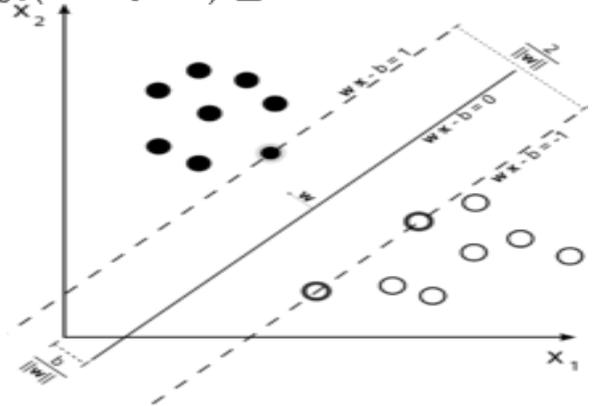


Figure 4: Maximum-margin hyperplane and margins for an SVM trained with samples from two classes. Samples on the margin are called the support vectors.

#### 3.4.2 Feed forward Backpropagation Neural Network

The feed forward, back-propagation architecture was developed in the early 1970's by several independent sources (Werbor; Parker; Rumelhart, Hinton and Williams) [20]. This independent co development was the result of a proliferation of articles and talks at various conferences which stimulated the entire industry[17].The backpropagation algorithm (Rumelhart and McClelland, 1986) is used in layered feed-forward ANNs. This means that the artificial neurons are organized in layers, and send their signals "forward", and then the errors are propagated backwards. The network receives inputs by neurons in the input layer, and the output of the network is given by the neurons on an output layer. There may be one or more intermediate hidden layers.

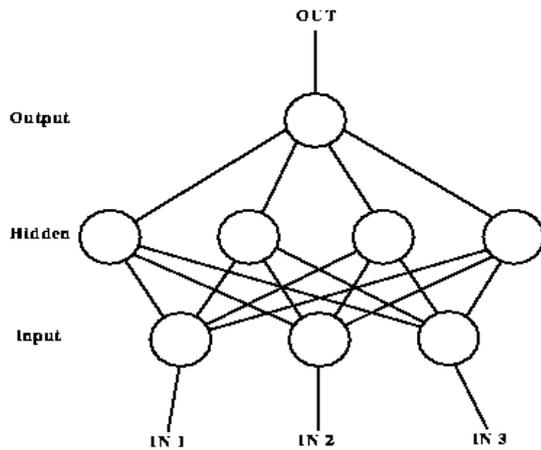


Figure 5: Neural network classifier architecture

The backpropagation algorithm uses supervised learning, which means that we provide the algorithm with examples of the inputs and outputs we want the network to compute, and then the error (difference between actual and expected results) is calculated. The idea of the backpropagation algorithm is to reduce this error, until the ANN learns the training data. The training begins with random weights, and the goal is to adjust them so that the Error will be minimal. [21].

### 3.5 Performance Evaluation

Three Statistical performance Measures are applied those are accuracy, sensitivity, positive predictive. These parameters are defined using 4 measures True Positive (TP), True Negative (TN), False Positive (FP), and False Negative (FN)[22].

**TP (True Positive):** Cardiomyopathy detection matches with decision of physician.

**TN (True Negative):** both classifier and physician advised absence of Cardiomyopathy.

**FP (False Positive):** system labels a healthy case as a Cardiomyopathy.

**FN(False Negative):** system labels a Cardiomyopathy as healthy

**Accuracy:** Accuracy is the ratio of number of correctly classified Samples, and is given by

$$\text{Accuracy} = \frac{TP+TN}{N} \times 100 \text{-----} (2)$$

Total number of cases are N

**Sensitivity:** Sensitivity refers to the rate of correctly classified positive. Sensitivity may be referred as a True Positive Rate. Sensitivity should be high for a classifier.

$$Se = \frac{TP}{TP+FN} \times 100 \text{-----} (3)$$

**Positive predictive:** Positive predictive is probability that disease is present when test is positive, which is by how much amount disease is correctly predicted.

$$Pp = \frac{TP}{TP+FP} \times 100 \text{-(4)}$$

## 4. RESULTS AND DISCUSSION

In SVM classification linear SVM is applied all the dataset is divided into Training dataset having total 46 samples and testing dataset having 17 samples. We are getting 94.11 % accuracy for Healthy and 60 % accuracy for cardiomyopathy

and Total 86 % classification accuracy is obtained. In ANN classification all the dataset are divided into 70% training, 15% Testing and 15% validation datasets. The network is having 8 input layers, 20 hidden layers and one output layer the system is trained using Levenberg-Marquardt algorithm the best classification accuracy is achieved up to 12 epochs. We are getting 98% accuracy for Healthy and 85% for Cardiomyopathy and total 95% classification accuracy. The results are described in Table 4.

Table 4: Classification Results.

Classifier	Accuracy	Healthy		Cardiomyopathy	
		Se	Pp	Se	Pp
Support Vector Machines	86.36	94.11	88.88	60	75
Feed forward backpropagation Neural Network	95.2	97.95	85.76	85.71	92.30

## 5. CONCLUSION AND FUTURE WORK

The proposed system works on the ECG signals taken from online PTB diagnostic ECG database total 63 subject's signals are obtained for Healthy and cardiomyopathy. Initially the signals are preprocessed for removal of high frequency noise and baseline correction. Adaptive thresholding method is then applied for the detection of P-QRS and T waves and the Intervals of the waves are calculated and taken as the features for the detection of cardiomyopathy and provided as Input to both classifiers SVM and ANN. Both the classifiers efficiently detect cardiomyopathy with good classification accuracy. Neural networks are the best classifier giving the highest classification accuracy 95%.wher as SVM giving 86 % accuracy.

In future the proposed system can be enhanced for more accurate results by using the advanced feature extraction techniques like DWT also the more classification methods can be applied to compare the performance. Finally the system can be updated to detect all four types of cardiomyopathy.

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